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Local limit theorem for densities of the additive component of a finite Markov Additive Process

Loïc HERVÉ and James LEDOUX *

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Abstract

In this paper, we are concerned with centered Markov Additive Processes $\{(X_t, Y_t)\}_{t \in \mathbb{T}}$ where the driving Markov process $\{X_t\}_{t \in \mathbb{T}}$ has a finite state space. Under suitable conditions, we provide a local limit theorem for the density of the absolutely continuous part of the probability distribution of $t^{-1/2}Y_t$ given X_0 . The rate of convergence and the moment condition are the expected ones with respect to the i.i.d case. An application to the joint distribution of local times of a finite jump process is sketched.

Keywords: Gaussian approximation, Local time, Spectral method, Markov random walk.

1 Introduction

When $\{X_k\}_{k \geq 1}$ is a sequence of centered independent and identically distributed (i.i.d.) real valued random variables such that $Y_n := \sum_{i=1}^n X_i$ has a bounded density for some n , it is well-known that the density f_n of $n^{-1/2}Y_n$ satisfies the following Local Limit Theorem (LLT)

$$\lim_{n \rightarrow +\infty} \sup_{y \in \mathbb{R}} |f_n(y) - \eta(y)| = 0,$$

where $\eta(\cdot)$ is the density of the Gaussian distribution $\mathcal{N}(0, \sigma^2)$, with $\sigma^2 := \mathbb{E}[X_1^2]$; see [Gne54] and [IL71, Fel71] for detailed discussions. If X_1 has a bounded density and a third moment, then the rate of the previous convergence is $O(n^{-1/2})$; see [SŠ65, Šah66, KZ98].

This paper extends the last result to centered Markov Additive Processes (MAP) $\{(X_t, Y_t)\}_{t \in \mathbb{T}}$ with state space $\mathbb{X} \times \mathbb{R}^d$, where $\mathbb{X} := \{1, \dots, N\}$ and $\mathbb{T} := \mathbb{N}$ or $\mathbb{T} := [0, \infty)$. Recall from [Asm03] that $\{(X_t, Y_t)\}_{t \in \mathbb{T}}$ is a Markov process on $\mathbb{X} \times \mathbb{R}^d$ with a transition semi-group, denoted by $\{Q_t\}_{t \in \mathbb{T}}$, which satisfies

$$\forall(k, y) \in \mathbb{X} \times \mathbb{R}^d, \forall(\ell, B) \in \mathbb{X} \times B(\mathbb{R}^d), \quad Q_t(k, y; \{\ell\} \times B) = Q_t(k, 0; \{\ell\} \times B - y). \quad (1)$$

The transition semi-group of the driving Markov process $\{X_t\}_{t \in \mathbb{T}}$ is denoted by $\{P_t\}_{t \in \mathbb{T}}$. The stochastic $N \times N$ -matrix $P := P_1$ is assumed to be irreducible and aperiodic. Moreover the mass of the singular part of the conditional probability distribution of $t^{-1/2}Y_t$ given $X_0 = k$ is supposed to converge exponentially fast to zero. Let $f_{k,t}(\cdot)$ be the density of the absolutely continuous part of this conditional distribution. Under a third moment condition on Y_t and some conditions on $f_{k,t}(\cdot)$ and its Fourier transform (see **(AC1)**-(**AC2**)), we prove

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in Theorem 2.2 that, for every $k \in \mathbb{X}$, the density $f_{k,t}(\cdot)$ satisfies essentially the following property:

$$\sup_{y \in \mathbb{R}^d} |f_{k,t}(y) - \eta_\Sigma(y)| = O(t^{-1/2})$$

where $\eta_\Sigma(\cdot)$ denotes the density of $\mathcal{N}(0, \Sigma)$. Matrix Σ is the asymptotic covariance provided by the Central Limit Theorem (CLT) and is assumed to be invertible. Our moment condition and rate of convergence are the expected ones with respect to the i.i.d. case. The proof of Theorem 2.2 is based on the spectral method (e.g. see [GH88, HH01] when $\mathbb{T} := \mathbb{N}$, and [FHL12] when $\mathbb{T} := [0, +\infty)$).

To the best of our knowledge, Theorem 2.2 is new. The known contribution to LLT for densities of additive components of MAP is in [RS08], where only the discrete time case is considered and exponential-type moment condition on Y_1 is assumed (the rate of convergence is not addressed). Note that, for discrete time, our Theorem 2.2 only requires a third moment condition on Y_1 . Moreover it is worth noticing that the probability distribution of Y_t is not assumed to be absolutely continuous with respect to the Lebesgue measure on \mathbb{R}^d . An application to the joint distribution of local times of a finite jump process is provided in Section 3. The proof of Theorem 2.2 is given in the last section. In order to save space, some details are reported in a companion paper which is referred to as [HL]¹.

This work was motivated by a question in relation with a self-attracting continuous time process called the Vertex Reinforced Jump Process (VRJP) recently investigated by [ST11]. This process is closely related to the Edge Reinforced Random Walk (ERRW) and was instrumental in the proof of the recurrence of the ERRW in all dimensions at strong reinforcement. In a paper in progress, [ST13] make a link between the VRJP and accurate pointwise large deviation for reversible Markov jump processes: it appears that the limit measure of the VRJP [ST11] is closely related to the first order of pointwise large deviations which are derived in [ST13] for continuous time Markov processes using the present local limit theorem and Remark 2.3.

Notations. Any vector $v \equiv (v_k) \in \mathbb{C}^N$ is considered as a row-vector and v^\top is the corresponding column-vector. The vector with all components equal to 1 is denoted by $\mathbf{1}$. The Euclidean scalar product and its associated norm on \mathbb{C}^N is denoted by $\langle \cdot, \cdot \rangle$ and $\|\cdot\|$ respectively. The set of $N \times N$ -matrices with complex entries is denoted by $\mathcal{M}_N(\mathbb{C})$. We use the following norm $\|\cdot\|_0$ on $\mathcal{M}_N(\mathbb{C})$:

$$\forall A \equiv (A_{k,\ell}) \in \mathcal{M}_N(\mathbb{C}), \quad \|A\|_0 := \max \{|A_{k,\ell}| : (k, \ell) \in \{1, \dots, N\}^2\}.$$

For any bounded positive measure ν on \mathbb{R}^d , we define its Fourier transform as:

$$\forall \zeta \in \mathbb{R}^d, \quad \widehat{\nu}(\zeta) := \int_{\mathbb{R}^d} e^{i\langle \zeta, y \rangle} d\nu(y).$$

Let $\mathcal{A} \equiv (\mathcal{A}_{k,\ell})$ be a $N \times N$ -matrix with entries in the set of bounded positive measures on \mathbb{R}^d . We set

$$\forall B \in \mathcal{B}(\mathbb{R}^d), \quad \mathcal{A}(1_B) := (\mathcal{A}_{k,\ell}(1_B)), \quad \forall \zeta \in \mathbb{R}^d, \quad \widehat{\mathcal{A}}(\zeta) := (\widehat{\mathcal{A}_{k,\ell}}(\zeta)). \quad (2)$$

2 The LLT for the density process

Let $\{(X_t, Z_t)\}_{t \in \mathbb{T}}$ be an MAP with state space $\mathbb{X} \times \mathbb{R}^d$, where $\mathbb{X} := \{1, \dots, N\}$ and the driving Markov process $\{X_t\}_{t \in \mathbb{T}}$ has transition semi-group $\{P_t\}_{t \in \mathbb{T}}$. We refer to [Asm03, Chap. XI]

¹Available at <http://arxiv.org/abs/1305.5644>

for the basic material on such MAPs. The conditional probability to $\{X_0 = k\}$ and its associated expectation are denoted by \mathbb{P}_k and \mathbb{E}_k respectively. Note that if $T : \mathbb{R}^d \rightarrow \mathbb{R}^m$ is a linear transformation, then $\{X_t, T(Z_t)\}_{t \in \mathbb{T}}$ is still a MAP on $\mathbb{X} \times \mathbb{R}^m$ (see Lem. C.1 in [HL]). We suppose that $\{X_t\}_{t \in \mathbb{T}}$ has a unique invariant probability measure π . Set $m := \mathbb{E}_\pi[Z_1] = \sum_k \pi(k) \mathbb{E}_k[Z_1] \in \mathbb{R}^d$. Consider the centered MAP $\{(X_t, Y_t)\}_{t \in \mathbb{T}}$ where $Y_t := Z_t - tm$. The two next assumptions are involved in both CLT and LLT below.

(I-A) : The stochastic $N \times N$ -matrix $P := P_1$ is irreducible and aperiodic.

(M α) : The family of r.v. $\{Y_v\}_{v \in (0,1] \cap \mathbb{T}}$ satisfies the uniform moment condition of order α :

$$M_\alpha := \max_{k \in \mathbb{X}} \sup_{v \in (0,1] \cap \mathbb{T}} \mathbb{E}_k[\|Y_v\|^\alpha] < \infty. \quad (3)$$

The next theorem provides a CLT for $t^{-1/2}Y_t$, proved when $d := 1$ in [KW64] for $\mathbb{T} = \mathbb{N}$ and in [FH67] for $\mathbb{T} = [0, \infty)$; see [FHL12] for ρ -mixing driving Markov processes.

Theorem 2.1 Under Assumptions **(I-A)** and **(M2)**, $\{t^{-1/2}Y_t\}_{t \in \mathbb{T}}$ converges in distribution to a d -dimensional Gaussian law $\mathcal{N}(0, \Sigma)$ when $t \rightarrow +\infty$.

Remark 2.1 When $\mathbb{T} = \mathbb{N}$, the moment condition reads as: $\mathbb{E}_\pi[\|Y_1\|^2] < \infty$. When $\mathbb{T} := [0, \infty)$, since \mathbb{X} is a finite set, we know that Condition **(M2)** on the family $\{Y_v\}_{v \in (0,1] \cap \mathbb{T}}$ is still equivalent to $\mathbb{E}_\pi[\|Y_1\|^2] < \infty$ [FH67]. Moreover, note that **(I-A)** is satisfied in the continuous time case provided that the generator of $\{X_t\}_{t \geq 0}$ is irreducible.

Now let us specify the notations and assumptions involved in our LLT. For any $t \in \mathbb{T}$ and $(k, \ell) \in \mathbb{X}^2$, we define the bounded positive measure $\mathcal{Y}_{k,\ell,t}$ on \mathbb{R}^d as follows (see (1)):

$$\forall B \in B(\mathbb{R}^d), \quad \mathcal{Y}_{k,\ell,t}(1_B) := \mathbb{P}_k\{X_t = \ell, Y_t \in B\} = Q_t(k, 0; \{\ell\} \times B). \quad (4)$$

Let ℓ_d denote the Lebesgue measure on \mathbb{R}^d . From the Lebesgue decomposition of $\mathcal{Y}_{k,\ell,t}$ w.r.t. ℓ_d , there are two bounded positive measures $\mathcal{G}_{k,\ell,t}$ and $\mu_{k,\ell,t}$ on \mathbb{R}^d such that

$$\forall B \in B(\mathbb{R}^d), \quad \mathcal{Y}_{k,\ell,t}(1_B) := \mathcal{G}_{k,\ell,t}(1_B) + \mu_{k,\ell,t}(1_B) \quad \text{where } \mathcal{G}_{k,\ell,t}(1_B) = \int_B g_{k,\ell,t}(y) dy \quad (5)$$

for some measurable function $g_{k,\ell,t} : \mathbb{R}^d \rightarrow [0, +\infty)$, and such that $\mu_{k,\ell,t}$ and ℓ_d are mutually singular. The measure $\mathcal{G}_{k,\ell,t}$ is called the absolutely continuous (a.c.) part of $\mathcal{Y}_{k,\ell,t}$ with associated density $g_{k,\ell,t}$. For any $t \in \mathbb{T}$, we introduce the following $N \times N$ -matrices with entries in the set of bounded positive measures on \mathbb{R}^d

$$\mathcal{Y}_t := (\mathcal{Y}_{k,\ell,t})_{(k,\ell) \in \mathbb{X}^2}, \quad \mathcal{G}_t := (\mathcal{G}_{k,\ell,t})_{(k,\ell) \in \mathbb{X}^2}, \quad \mathcal{M}_t := (\mu_{k,\ell,t})_{(k,\ell) \in \mathbb{X}^2}, \quad (6a)$$

and for every $y \in \mathbb{R}^d$, we define the following real $N \times N$ -matrix:

$$G_t(y) := (g_{k,\ell,t}(y))_{(k,\ell) \in \mathbb{X}^2}. \quad (6b)$$

Then the component-wise equalities (5) read as follows in a matrix form: for any $t \in \mathbb{T}$

$$\forall B \in B(\mathbb{R}^d), \quad \mathcal{Y}_t(1_B) = \mathcal{G}_t(1_B) + \mathcal{M}_t(1_B) = \int_B G_t(y) dy + \mathcal{M}_t(1_B). \quad (6c)$$

The assumptions on the a.c. part \mathcal{G}_t and the singular part \mathcal{M}_t of \mathcal{Y}_t are the following ones.

AC1 : There exist $c > 0$ and $\rho \in (0, 1)$ such that

$$\forall t > 0, \quad \|\mathcal{M}_t(1_{\mathbb{R}^d})\|_0 \leq c\rho^t \quad (7)$$

and there exists $t_0 > 0$ such that $\rho^{t_0} \max(2, cN) \leq 1/4$ and

$$\Gamma_{t_0}(\zeta) := \sup_{w \in [t_0, 2t_0)} \|\widehat{G}_w(\zeta)\|_0 \longrightarrow 0 \quad \text{when } \|\zeta\| \rightarrow +\infty. \quad (8)$$

AC2 : For any $t > 0$, there exists an open convex subset \mathcal{D}_t of \mathbb{R}^d such that G_t vanishes on $\mathbb{R}^d \setminus \overline{\mathcal{D}_t}$, where $\overline{\mathcal{D}_t}$ denotes the closure of \mathcal{D}_t . Moreover G_t is continuous on $\overline{\mathcal{D}_t}$ and differentiable on \mathcal{D}_t , with in addition

$$\begin{aligned} \sup_{t>0} \sup_{y \in \overline{\mathcal{D}_t}} \|G_t(y)\|_0 &< \infty & \sup_{y \in \partial \mathcal{D}_t} \|G_t(y)\|_0 &= O(t^{-(d+1)/2}) \quad \text{where } \partial \mathcal{D}_t := \overline{\mathcal{D}_t} \setminus \mathcal{D}_t \\ j = 1, \dots, d : \quad \sup_{t>0} \sup_{y \in \mathcal{D}_t} \left\| \frac{\partial G_t}{\partial y_j}(y) \right\|_0 &< \infty. \end{aligned}$$

The next theorem is the main result of the paper.

Theorem 2.2 Let $\{(X_t, Y_t)\}_{t \in \mathbb{T}}$ be a centered MAP satisfying Assumptions **(I-A)**, **(M3)**, **(AC1)**-**(AC2)**. Moreover assume that the matrix Σ associated with the CLT in Theorem 2.1 is invertible. Then, for every $k \in \mathbb{X}$, the density $f_{k,t}(\cdot)$ of the a.c. part of the probability distribution of $t^{-1/2}Y_t$ under \mathbb{P}_k satisfies the following property:

$$\sup_{y \in \mathbb{R}^d} |f_{k,t}(y) - \eta_\Sigma(y)| \leq O(t^{-1/2}) + O\left(\sup_{y \notin \mathcal{D}_t} \eta_\Sigma(t^{-1/2}y)\right) \quad (10)$$

where $\eta_\Sigma(\cdot)$ denotes the density of $\mathcal{N}(0, \Sigma)$.

Remark 2.2 (Non-lattice condition) To derive a LLT, it is standard to assume some non-lattice condition which writes for the MAP $\{(X_t, Y_t)\}_{t \in \mathbb{T}}$ as (e.g see [FHL12]):

(NL) : there is no triplet (a, F, θ) with $a \in \mathbb{R}^d$, F a closed subgroup of \mathbb{R}^d , $F \neq \mathbb{R}^d$ and $\beta : \mathbb{X} \rightarrow \mathbb{R}^d$ such that: $\forall k \in \mathbb{X}, \quad Y_1 + \beta(X_1) - \beta(k) \in a + F \quad \mathbb{P}_k\text{-a.s.}$

Actually Condition **(NL)** holds under **(I-A)** and **(AC1)** (see the proof of Lemma 4.2). Moreover, when $\{Y_t\}_{t \geq 0}$ is an additive functional of $\{X_t\}_{t \geq 0}$, the matrix Σ in the CLT for $\{t^{-1/2}Y_t\}_{t > 0}$ is invertible under **(NL)**. Indeed, if not, there is $\zeta_0 \in \mathbb{R}^N$ such that $\langle \zeta_0, \Sigma \zeta_0 \rangle = 0$. Thus, $\{t^{-1/2}\langle \zeta_0, Y_t \rangle\}_{t > 0}$ converges in distribution to 0 (which is a degenerate Gaussian random variable). It follows from [HS70, Th. 2.2] that, for every $k \in \mathbb{X}$, $\langle \zeta_0, Y_1 \rangle = b(X_1) - b(k) \quad \mathbb{P}_k\text{-a.s.}$ for some $b : \mathbb{X} \rightarrow \mathbb{R}$. But this is impossible under **(NL)**.

Remark 2.3 (A uniform LLT with respect to transition matrix P) Let \mathcal{P} denote the set of irreducible and aperiodic stochastic $N \times N$ -matrices, equipped with the topology associated with (for instance) the distance $d(P, P') := \|P - P'\|_0$ ($P, P' \in \mathcal{P}$). Assume that there is some compact subset \mathcal{P}_0 of \mathcal{P} such that $P := P_1 \in \mathcal{P}_0$, where $\{P_t\}_{t \in \mathbb{T}}$ is the transition semi-group of $\{X_t\}_{t \in \mathbb{T}}$. To keep in mind the dependence on P , the positive measures $\mathcal{Y}_{k,\ell,t}$ in (4) and the matrix Σ in Theorem 2.1 are denoted by $\mathcal{Y}_{k,\ell,1}^P$ and Σ^P respectively. Finally let (\mathfrak{M}, d_{TV}) be the space of bounded positive measures on \mathbb{R}^d equipped with the total variation distance d_{TV} (i.e. $\forall (\mu, \mu') \in \mathfrak{M}^2$, $d_{TV}(\mu, \mu') := \sup\{|\mu(f) - \mu'(f)|, |f| \leq 1\}$). Let us assume that the centered MAP $\{(X_t, Y_t)\}_{t \in \mathbb{T}}$ satisfies the following conditions:

U1 : For any $(k, \ell) \in \mathbb{X}^2$, the map $P \mapsto \mathcal{Y}_{k, \ell, 1}^P$ is continuous from (\mathcal{P}_0, d) into (\mathfrak{M}, d_{TV}) .

U2 : There exist positive constants α and β such that

$$\forall P \in \mathcal{P}_0, \forall \zeta \in \mathbb{R}^d, \quad \alpha \|\zeta\|^2 \leq \langle \zeta, \Sigma^P \zeta \rangle \leq \beta \|\zeta\|^2. \quad (11)$$

U3 : The conditions **(M3)**, **(AC1)** and **(AC2)** hold uniformly in $P \in \mathcal{P}_0$.

Then, under Assumptions **(U1)**-**(U3)**, for every $k \in \mathbb{X}$, the density $f_{k, t}^P(\cdot)$ of the a.c. part of the probability distribution of $t^{-1/2}Y_t$ under \mathbb{P}_k satisfies the following property when $t \rightarrow +\infty$:

$$\sup_{P \in \mathcal{P}_0} \sup_{y \in \mathbb{R}^d} |f_{k, t}^P(y) - \eta_{\Sigma^P}(y)| = O(t^{-1/2}) + O\left(\sup_{P \in \mathcal{P}_0} \sup_{y \notin \mathcal{D}_t} \eta_{\Sigma^P}(t^{-1/2}y)\right).$$

This result follows from a suitable adaptation of the proof of Theorem 2.2 (see Remark 4.1).

3 Application to the local times of a jump process

Let $\{X_t\}_{t \geq 0}$ be a Markov jump process with finite state space $\mathbb{X} := \{1, \dots, N\}$ and generator G . Its transition semi-group is given by: $\forall t \geq 0, P_t := e^{tG}$. The local time $L_t(i)$ associated with state $i \in \mathbb{X}$, or the sojourn time in state i on the interval $[0, t]$, is defined by

$$\forall t \geq 0, \quad L_t(i) := \int_0^t 1_{\{X_s = i\}} ds.$$

It is well known that $L_t(i)$ is an additive functional of $\{X_t\}_{t \geq 0}$ and that $\{(X_t, L_t(i))\}_{t \geq 0}$ is an MAP. Consider the MAP $\{(X_t, L_t)\}_{t \geq 0}$ where L_t is the random vector of the local times $L_t := (L_t(1), \dots, L_t(N))$. Note that, for all $t > 0$, we have $\langle L_t, \mathbf{1} \rangle = t$, that is L_t is \mathcal{S}_t -valued where

$$\mathcal{S}_t := \{y \in [0, +\infty)^N : \langle y, \mathbf{1} \rangle = t\}.$$

Assume that G is irreducible. Then Condition **(I-A)** holds true and $\{X_t\}_{t \geq 0}$ has a unique invariant probability measure π . Set $m = (m_1, \dots, m_N) := \mathbb{E}_\pi[L_1]$, and define the $\mathcal{S}_t^{(0)}$ -valued centered r.v.

$$Y_t = L_t - tm$$

where $\mathcal{S}_t^{(0)} := T_{-tm}(\mathcal{S}_t)$ with the translation T_{-tm} by vector $-tm$ in \mathbb{R}^N . Note that $\mathcal{S}_t^{(0)}$ is a subset of the hyperplane H of \mathbb{R}^N defined by

$$H := \{y \in \mathbb{R}^N : \langle y, \mathbf{1} \rangle = 0\}. \quad (12)$$

We denote by Λ the bijective map from H into \mathbb{R}^{N-1} defined by $\Lambda(y) := (y_1, \dots, y_{N-1})$ for $y := (y_1, \dots, y_N) \in H$. Recall that ℓ_{N-1} denotes the Lebesgue measure on \mathbb{R}^{N-1} . Let ν be the measure defined on $(H, B(H))$ as the image measure of ℓ_{N-1} under Λ^{-1} , where $B(H)$ stands for the Borelian σ -algebra on H . Finally, under the probability measure \mathbb{P}_k , $f_{k, t}$ is the density of the a.c. part of the probability distribution of $t^{-1/2}Y_t = t^{-1/2}(L_t - tm)$ with respect to the measure ν on H .

Proposition 3.1 *If G and the sub-generators $G_{i^c i^c} := (G(k, \ell))_{k, \ell \in \{i\}^c}$, $i = 1, \dots, N$ are irreducible, then there exists a definite-positive $(N-1) \times (N-1)$ -matrix Σ such that:*

$$\forall k \in \mathbb{X}, \quad \sup_{h \in H} |f_{k, t}(h) - (2\pi)^{-(N-1)/2} (\det \Sigma)^{-1/2} e^{-\frac{1}{2} \langle \Lambda h^\top, \Sigma \Lambda h^\top \rangle}| = O(t^{-1/2}).$$

Proof. Let Y'_t be the $(N-1)$ -dimensional random vector $\Lambda(Y_t) = (Y_t(1), \dots, Y_t(N-1))$. Then $\{(X_t, Y'_t)\}_{t \geq 0}$ is a $\mathbb{X} \times \overline{\mathcal{D}}_t$ -valued MAP where \mathcal{D}_t is the following open convex of \mathbb{R}^{N-1}

$$\mathcal{D}_t := \{y' \in \mathbb{R}^{N-1} : j = 1, \dots, N-1, y'_j \in (-m_j t, (1-m_j)t), \langle y', \mathbf{1} \rangle < m_N t\}. \quad (13)$$

Using the assumptions on G and results of [Ser00], the MAP $\{(X_t, Y'_t)\}_{t \geq 0}$ satisfies the conditions **(M3)** and **(AC1)**-**(AC2)** (see details in Section 5 of [HL]). The matrix Σ in the CLT for $\{t^{-1/2}Y'_t\}_{t > 0}$ is invertible from Remark 2.2 and $O(\sup_{y \notin \mathcal{D}_t} \eta_\Sigma(t^{-1/2}y)) = O(t^{-1/2})$ from (13). Thus Theorem 2.2 gives that

$$\forall k \in \mathbb{X}, \quad \sup_{y' \in \mathbb{R}^{N-1}} |f'_{k,t}(y') - (2\pi)^{-(N-1)/2} (\det \Sigma)^{-1/2} e^{-\frac{1}{2} \langle y'^\top, \Sigma y'^\top \rangle}| = O(t^{-1/2}),$$

where $f'_{k,t}$ is the density of the a.c. part of the probability distribution of $t^{-1/2}Y'_t$. Finally, using the bijection Λ , it is easily seen that the density $f_{k,t}$ is given by: $\forall h \in H, f_{k,t}(h) := f'_{k,t}(\Lambda h)$. This gives the claimed result. \square

Remark 3.1 *First, the assumption on the sub-generators of G in Proposition 3.1 is used to obtain the geometric convergence to 0 of the mass of the singular part of the probability distribution of Y'_t required in **(AC1)** (see Section 5 of [HL] for details). Second, it can be seen from [Pin91] that Σ is the definite-positive matrix*

$$\forall (i, j) \in \{1, \dots, N-1\}^2, \quad \Sigma_{i,j} = \pi_i C_{i,j} + \pi_j C_{j,i} \quad \text{where} \quad C := - \int_0^\infty (e^{tG} - \mathbf{1}^\top \pi) dt.$$

4 Fourier analysis and proof of Theorem 2.2

Assume that the conditions of Theorem 2.2 hold. Note that $\{(X_t, \Sigma^{-1/2}Y_t)\}_{t \in \mathbb{T}}$ is an MAP with the identity matrix I as asymptotic covariance matrix. It still satisfies the assumptions of Theorem 2.2 since the associated bounded positive measures of (5) are the image measure of $\mathcal{G}_{k,\ell,t}$ and $\mathcal{M}_{k,\ell,t}$ under the bijective linear map $\Sigma^{-1/2}$. Thus, we only have to prove Theorem 2.2 when $\Sigma = I$. This proof involves Fourier analysis as in the i.i.d. case. In our Markov context, this study is based on the semi-group property of the matrix family $\{\hat{\mathcal{Y}}_t(\zeta)\}_{t \in \mathbb{T}}$ for every $\zeta \in \mathbb{R}^d$ which allows us to analyze the characteristic function of Y_t . In the next subsection, we provide a collection of lemmas which highlights the connections between the assumptions of Section 2 and the behavior of this semi-group. This is the basic material for the derivation of Theorem 2.2 in Subsection 4.2.

Let us denote the integer part of any $t \in \mathbb{T}$ by $\lfloor t \rfloor$. Using the notation of (5), the bounded positive measure $\sum_{\ell=1}^N \mathcal{G}_{k,\ell,t}$ with density $g_{k,t} := \sum_{\ell=1}^N g_{k,\ell,t}$ is the a.c. part of the probability distribution of Y_t under \mathbb{P}_k , while $\mu_{k,t} := \sum_{\ell=1}^N \mu_{k,\ell,t}$ is its singular part. That is, we have for any $k \in \mathbb{X}$ and $t > 0$:

$$\forall B \in \mathcal{B}(\mathbb{R}^d), \quad \mathbb{P}_k\{Y_t \in B\} = \int_B g_{k,t}(y) dy + \mu_{k,t}(1_B). \quad (14)$$

4.1 Semigroup of Fourier matrices and basic lemmas

The bounded positive measure \mathcal{Y}_t is defined in (6a) and its Fourier transform $\hat{\mathcal{Y}}_t$ is (see (2)):

$$\forall (t, \zeta) \in \mathbb{T} \times \mathbb{R}^d, \quad \forall (k, \ell) \in \mathbb{X}^2, \quad (\hat{\mathcal{Y}}_t(\zeta))_{k,\ell} = \mathbb{E}_k[1_{\{X_t=\ell\}} e^{i \langle \zeta, Y_t \rangle}]. \quad (15)$$

Note that $\widehat{\mathcal{Y}}_t(0) = P_t$. From the additivity of the second component Y_t , we know that $\{\widehat{\mathcal{Y}}_t(\zeta)\}_{t \in \mathbb{T}, \zeta \in \mathbb{R}^d}$ is a semi-group of matrices (e.g. see [FHL12] for details), that is

$$\forall \zeta \in \mathbb{R}^d, \forall (s, t) \in \mathbb{T}^2, \quad \widehat{\mathcal{Y}}_{t+s}(\zeta) = \widehat{\mathcal{Y}}_t(\zeta) \widehat{\mathcal{Y}}_s(\zeta). \quad (\text{SG})$$

In particular the following property holds true

$$\forall \zeta \in \mathbb{R}^d, \forall n \in \mathbb{N}, \quad \widehat{\mathcal{Y}}_n(\zeta) := (\mathbb{E}_k[1_{\{X_n=\ell\}} e^{i\langle \zeta, Y_n \rangle}])_{(k, \ell) \in \mathbb{X}^2} = \widehat{\mathcal{Y}}_1(\zeta)^n. \quad (16)$$

For every $k \in \mathbb{X}$ and $t \in \mathbb{T}$, we denote by $\phi_{k,t}$ the characteristic function of Y_t under \mathbb{P}_k :

$$\forall \zeta \in \mathbb{R}^d, \quad \phi_{k,t}(\zeta) := \mathbb{E}_k[e^{i\langle \zeta, Y_t \rangle}] = e_k \widehat{\mathcal{Y}}_t(\zeta) \mathbf{1}^\top, \quad (17)$$

where e_k is the k -th vector of the canonical basis of \mathbb{R}^d .

Under Conditions **(I-A)** and **(M2)**, the spectral method provides the CLT for $\{(t^{-1/2} Y_t)\}_{t \in \mathbb{T}}$. To extend this CLT to a local limit theorem, a precise control of the characteristic function $\phi_{k,t}$ is needed. This is the purpose of the next three lemmas, in which the semi-group property (SG) plays an important role. The first lemma, which is the central part in the proof of Theorem 2.2, provides the control of $\phi_{k,t}$ on a ball $B(0, \delta) := \{\zeta \in \mathbb{R}^d : \|\zeta\| < \delta\}$ for some $\delta > 0$. The second one is on the control of $\phi_{k,t}$ on the annulus $\{\zeta \in \mathbb{R}^d : \delta \leq \|\zeta\| \leq A\}$ for any $A > \delta$. Finally the third lemma focuses on the Fourier transform of the density $g_{k,\ell,t}$ of the a.c. part $\mathcal{G}_{k,\ell,t}$ of $\mathcal{Y}_{k,\ell,t}$ in (5) on the domain $\{\zeta \in \mathbb{R}^d : \|\zeta\| \geq A\}$ for some $A > 0$.

Lemma 4.1 *Under Assumptions **(I-A)** and **(M3)**, there exists a real number $\delta > 0$ such that, for all $\zeta \in B(0, \delta)$, the characteristic function of Y_t satisfies*

$$\forall k \in \mathbb{X}, \forall t \in \mathbb{T}, \quad \phi_{k,t}(\zeta) = \lambda(\zeta)^{\lfloor t \rfloor} L_{k,t}(\zeta) + R_{k,t}(\zeta), \quad (18)$$

with \mathbb{C} -valued functions $\lambda(\cdot)$, $L_{k,t}(\cdot)$ and $R_{k,t}(\cdot)$ on $B(0, \delta)$ satisfying the next properties for $t \in \mathbb{T}$, $\zeta \in \mathbb{R}^d$ and $k \in \mathbb{X}$:

$$\|\zeta\| < \delta \quad \Rightarrow \quad \lambda(\zeta) = 1 - \|\zeta\|^2/2 + O(\|\zeta\|^3) \quad (19a)$$

$$t \geq 2, \quad \|t^{-1/2} \zeta\| < \delta \quad \Rightarrow \quad |\lambda(t^{-1/2} \zeta)^{\lfloor t \rfloor} - e^{-\|\zeta\|^2/2}| \leq t^{-1/2} C (1 + \|\zeta\|^3) e^{-\|\zeta\|^2/8} \quad (19b)$$

$$\|\zeta\| < \delta \quad \Rightarrow \quad |L_{k,t}(\zeta) - 1| \leq C \|\zeta\| \quad (19c)$$

$$\exists r \in (0, 1), \quad \sup\{|R_{k,t}(\zeta)| : \|\zeta\| \leq \delta\} \leq C r^{\lfloor t \rfloor} \quad (19d)$$

where the constant $C > 0$ in (19b)-(19d) only depends on δ and on M_3 in **(M3)** (see (3)).

Proof. Assumption **(I-A)** ensures that $\widehat{\mathcal{Y}}_1(0) = P_1 = \Pi + N$, where $\Pi := (\Pi_{i,j})_{(i,j) \in \mathbb{X}}$ is the rank-one matrix defined by $\Pi_{i,j} := \pi(i)$ and $N \in \mathcal{M}_N(\mathbb{C})$ is such that $\|N^n\|_0 = O(\kappa^n)$ for some $\kappa \in (0, 1)$. Moreover note that the map $\zeta \mapsto \widehat{\mathcal{Y}}_1(\zeta)$ is thrice-differentiable from \mathbb{R}^d into $\mathcal{M}_N(\mathbb{C})$ due to (15) and **(M3)**. Then the standard perturbation theory shows that, for any $r \in (\kappa, 1)$, there exists $\delta \equiv \delta(r) > 0$ such that, for all $\zeta \in B(0, \delta)$, $\widehat{\mathcal{Y}}_1(\zeta)^n = \lambda(\zeta)^n \Pi(\zeta) + N(\zeta)^n$ where $\lambda(\zeta)$ is the dominating eigenvalue of $\widehat{\mathcal{Y}}_1(\zeta)$, $\Pi(\zeta)$ is the associated rank-one eigenprojection, and $N(\zeta) \in \mathcal{M}_N(\mathbb{C})$ is such that $\|N(\zeta)^n\|_0 = O(r^n)$. Moreover the maps $\lambda(\cdot)$, $\Pi(\cdot)$, and $N(\cdot)$ are thrice-differentiable on $B(0, \delta)$. These properties and (17) provide all the conclusions of Lemma 4.1 for $t \in \mathbb{N}$ (see [HH01, Prop. VI.2] for details). For Properties (19a), (19c) and (19d), the passage to the continuous-time case $\mathbb{T} := [0, +\infty)$ can be easily derived from $\phi_{k,t}(\zeta) = e_k \widehat{\mathcal{Y}}_1(\zeta)^{\lfloor t \rfloor} \widehat{\mathcal{Y}}_v(\zeta) \mathbf{1}^\top$ with $v := t - \lfloor t \rfloor \in [0, 1)$ and the fact that $\zeta \mapsto \widehat{\mathcal{Y}}_v(\zeta)$ is thrice-differentiable from \mathbb{R}^d into $\mathcal{M}_N(\mathbb{C})$ with partial derivatives uniformly bounded in $v \in (0, 1]$. See [FHL12, Prop. 4.4] for details. For (19b), the passage to $\mathbb{T} := [0, +\infty)$ is an easy extension of Inequality (v) of [HH01, Prop. VI.2]) (see details in [HL]). \square

Lemma 4.2 Assume that Conditions **(I-A)**, **(AC1)** and **(M α)** for some $\alpha > 0$ hold. Let δ, A be any real numbers such that $0 < \delta < A$. There exist constants $D \equiv D(\delta, A) > 0$ and $\tau \equiv \tau(\delta, A) \in (0, 1)$ such that

$$\forall k \in \mathbb{X}, \forall t \in \mathbb{T}, \quad \sup \{ |\phi_{k,t}(\zeta)| : \delta \leq \|\zeta\| \leq A \} \leq D \tau^{\lfloor t \rfloor}. \quad (20)$$

Proof. Lemma 4.2 can be classically derived from the spectral method under the non-lattice condition **(NL)** of Remark 2.2 and moment condition **(M α)** for some $\alpha > 0$ [FHL12, p. 412] (see also the proof of Lem. 6.2 in [HL] for details). Therefore, it is enough to prove that $\{Y_t\}_{t \in \mathbb{T}}$ satisfies **(NL)** under Assumptions **(I-A)** and **(AC1)**.

For t large enough, the function $g_{k,t}$ is nonzero in the Lebesgue space $\mathbb{L}^1(\mathbb{R}^d)$ since the mass $\mu_{k,t}(1_{\mathbb{R}^d})$ goes to 0 when $t \rightarrow +\infty$ from Assumption **(AC1)**. Let us fix $k \in \mathbb{X}$ and some integer $q \in \mathbb{N}^*$ such that $g_{k,q} \neq 0$ in $\mathbb{L}^1(\mathbb{R}^d)$. Assume that $\{Y_t\}_{t \in \mathbb{T}}$ is lattice. Then there exist $a \in \mathbb{R}^d$, $\theta : \mathbb{X} \rightarrow \mathbb{R}^d$ and a closed subgroup F in \mathbb{R}^d , $F \neq \mathbb{R}^d$, such that

$$Y_q + \theta(X_q) - \theta(k) \in a + F \quad \mathbb{P}_k - \text{a.s.}^2.$$

For any $\ell \in \mathbb{X}$, define the following subset of \mathbb{R}^d : $F_{k,\ell} := \theta(k) - \theta(\ell) + a + F$. Note that $\ell_d(F_{k,\ell}) = 0$ since F is a proper closed subgroup of \mathbb{R}^d . Since $\mu_{k,\ell,q}$ and ℓ_d are singular, there is $E_{k,\ell} \in B(\mathbb{R}^d)$ such that $\ell_d(E_{k,\ell}) = 0$ and $\mu_{k,\ell,q}(B) = \mu_{k,\ell,q}(B \cap E_{k,\ell})$ for any $B \in B(\mathbb{R}^d)$. Set $E_k := \cup_{j=1}^N E_{k,j}$ and $E_k^c := \mathbb{R}^d \setminus E_k$. Observe that $\mu_{k,\ell,q}(E_k^c) = 0$. We obtain

$$\begin{aligned} \forall B \in B(\mathbb{R}^d), \quad \mathbb{P}_k \{ Y_q \in B \cap E_k^c \} &= \sum_{\ell=1}^N \mathbb{P}_k \{ X_q = \ell, Y_q \in F_{k,\ell} \cap B \cap E_k^c \} \\ &= \sum_{\ell=1}^N \int_{F_{k,\ell} \cap B \cap E_k^c} g_{k,\ell,q}(y) dy + \sum_{\ell=1}^N \mu_{k,\ell,q}(1_{F_{k,\ell} \cap B \cap E_k^c}) = 0. \end{aligned}$$

On the other hand, we have using (14)

$$\forall B \in B(\mathbb{R}^d), \quad \mathbb{P}_k \{ Y_q \in B \cap E_k^c \} = \int_{B \cap E_k^c} g_{k,q}(y) dy + \mu_{k,q}(1_{B \cap E_k^c}) = \int_{B \cap E_k^c} g_{k,q}(y) dy.$$

From these equalities and $\ell_d(E_k) = 0$, it follows that $\int_B g_{k,q}(y) dy = \int_{B \cap E_k^c} g_{k,q}(y) dy = 0$. But this is impossible since $g_{k,q} \neq 0$ in $\mathbb{L}^1(\mathbb{R}^d)$. \square

Lemma 4.3 Under Condition **(AC1)**, there exist positive constants A and C such that

$$|\zeta| \geq A \implies \forall (k, \ell) \in \mathbb{X}^2, \forall t \in [t_0, +\infty), \quad |\widehat{g}_{k,\ell,t}(\zeta)| \leq C \frac{t}{2^{t/t_0}}.$$

Proof of Lemma 4.3. Let us introduce the following matrix norm: $\forall A \in \mathcal{M}_N(\mathbb{C}), \quad \|A\|_\infty := \sup_{\|v\|_0=1} \|Av^\top\|_0$. It is easily seen that

$$\forall A \in \mathcal{M}_N(\mathbb{C}), \quad \|A\|_0 \leq \|A\|_\infty \leq N \|A\|_0. \quad (21)$$

²The lattice condition is classically expressed under \mathbb{P}_π with $q = 1$ by using the spectral method. However this condition can be similarly expressed with Y_q for any $q \in \mathbb{N}^*$. Moreover it can be stated under \mathbb{P}_k for every $k \in \mathbb{X}$ since \mathbb{X} is finite and P is irreducible and aperiodic from **(I-A)**.

Let $(t, \zeta) \in \mathbb{T} \times \mathbb{R}^d$. From (6c) we have $\widehat{\mathcal{Y}}_t(\zeta) = \widehat{\mathcal{G}}_t(\zeta) + \widehat{\mathcal{M}}_t(\zeta)$ and the semi-group property (SG) gives for any $(s, t) \in \mathbb{T}^2$ and $\zeta \in \mathbb{R}^d$:

$$\begin{aligned}\widehat{\mathcal{Y}}_{t+s}(\zeta) &:= \widehat{\mathcal{G}}_{t+s}(\zeta) + \widehat{\mathcal{M}}_{t+s}(\zeta) = \widehat{\mathcal{Y}}_t(\zeta) \widehat{\mathcal{Y}}_s(\zeta) \\ &= (\widehat{\mathcal{G}}_t(\zeta) + \widehat{\mathcal{M}}_t(\zeta))(\widehat{\mathcal{G}}_s(\zeta) + \widehat{\mathcal{M}}_s(\zeta)) \\ &= \widehat{\mathcal{G}}_t(\zeta)\widehat{\mathcal{G}}_s(\zeta) + \widehat{\mathcal{G}}_t(\zeta)\widehat{\mathcal{M}}_s(\zeta) + \widehat{\mathcal{M}}_t(\zeta)\widehat{\mathcal{G}}_s(\zeta) + \widehat{\mathcal{M}}_t(\zeta)\widehat{\mathcal{M}}_s(\zeta).\end{aligned}$$

Thus

$$\widehat{\mathcal{G}}_{t+s}(\zeta) = \widehat{\mathcal{G}}_t(\zeta)\widehat{\mathcal{G}}_s(\zeta) + \widehat{\mathcal{G}}_t(\zeta)\widehat{\mathcal{M}}_s(\zeta) + \widehat{\mathcal{M}}_t(\zeta)\widehat{\mathcal{G}}_s(\zeta) + \widehat{\mathcal{M}}_t(\zeta)\widehat{\mathcal{M}}_s(\zeta) - \widehat{\mathcal{M}}_{t+s}(\zeta).$$

Then, we obtain using the matrix norm $\|\cdot\|_\infty$:

$$\begin{aligned}\|\widehat{\mathcal{G}}_{t+s}(\zeta)\|_\infty &\leq \|\widehat{\mathcal{G}}_t(\zeta)\|_\infty \|\widehat{\mathcal{G}}_s(\zeta)\|_\infty + \|\widehat{\mathcal{G}}_t(\zeta)\|_\infty \|\widehat{\mathcal{M}}_s(\zeta)\|_\infty + \|\widehat{\mathcal{M}}_t(\zeta)\|_\infty \|\widehat{\mathcal{G}}_s(\zeta)\|_\infty \\ &\quad + \|\widehat{\mathcal{M}}_t(\zeta)\|_\infty \|\widehat{\mathcal{M}}_s(\zeta)\|_\infty + \|\widehat{\mathcal{M}}_{t+s}(\zeta)\|_\infty.\end{aligned}\quad (22)$$

Moreover we deduce from (7) and (21) with ρ and c given in Condition **(AC1)**, that:

$$\forall u > 0, \forall \zeta \in \mathbb{R}^d, \quad \|\widehat{\mathcal{M}}_u(\zeta)\|_\infty \leq Nc\rho^u. \quad (23)$$

We set, with t_0 given in **(AC1)**,

$$K := \max(2, Nc) \quad \text{and} \quad \Gamma(\zeta) = \max(2\rho^{t_0}, \sup_{w \in [t_0, 2t_0)} \|\widehat{\mathcal{G}}_w(\zeta)\|_\infty).$$

Let us prove by induction that the following inequality holds for any $n \in \mathbb{N}^*$ and $v \in [0, t_0)$:

$$\forall \zeta \in \mathbb{R}^d, \quad \|\widehat{\mathcal{G}}_{nt_0+v}(\zeta)\|_\infty \leq [1 + (1 + K)(n - 1)] \Gamma(\zeta) (\Gamma(\zeta) + K\rho^{t_0})^{n-1}. \quad (24)$$

First, (24) is obvious for $n := 1$. Second assume that (24) holds for some $n \in \mathbb{N}^*$. Then, using (22) with $t := t_0 + v$ and $s := nt_0$, it follows from (23) and the definition of $\Gamma(\cdot)$ that

$$\begin{aligned}\|\widehat{\mathcal{G}}_{(n+1)t_0+v}(\zeta)\|_\infty &= \|\widehat{\mathcal{G}}_{t_0+v+nt_0}(\zeta)\|_\infty \\ &\leq \Gamma(\zeta) \|\widehat{\mathcal{G}}_{nt_0}(\zeta)\|_\infty + \Gamma(\zeta) K \rho^{nt_0} + K \rho^{t_0} \|\widehat{\mathcal{G}}_{nt_0}(\zeta)\|_\infty + 2K^2 \rho^{t_0+nt_0} \\ &\leq (\Gamma(\zeta) + K\rho^{t_0}) \|\widehat{\mathcal{G}}_{nt_0}(\zeta)\|_\infty + \Gamma(\zeta) K \rho^{nt_0} + 2K^2 \rho^{t_0+nt_0} \\ (\text{by induction}) &\leq [1 + (1 + K)(n - 1)] \Gamma(\zeta) (\Gamma(\zeta) + K\rho^{t_0})^n + \Gamma(\zeta) K \rho^{nt_0} + 2K^2 \rho^{t_0+nt_0}.\end{aligned}$$

Next, using $K\rho^{nt_0} \leq (K\rho^{t_0})^n \leq (\Gamma(\zeta) + K\rho^{t_0})^n$ and $2\rho^{t_0} \leq \Gamma(\zeta)$, we obtain

$$\begin{aligned}\Gamma(\zeta) K \rho^{nt_0} + 2K^2 \rho^{t_0+nt_0} &\leq \Gamma(\zeta) (\Gamma(\zeta) + K\rho^{t_0})^n + 2K\rho^{t_0} (\Gamma(\zeta) + K\rho^{t_0})^n \\ &\leq (1 + K) \Gamma(\zeta) (\Gamma(\zeta) + K\rho^{t_0})^n.\end{aligned}$$

Hence

$$\|\widehat{\mathcal{G}}_{(n+1)t_0+v}(\zeta)\|_\infty \leq [1 + (1 + K)n] \Gamma(\zeta) (\Gamma(\zeta) + K\rho^{t_0})^n.$$

The induction is complete and proves (24).

Now let $t \in [t_0, +\infty)$. By writing $t = nt_0 + v$ with $n \in \mathbb{N}^*$ and $v \in [0, t_0)$, Lemma 4.3 follows from (24) and Condition **(AC1)**. Indeed (21) gives $\Gamma(\zeta) \leq \max(2\rho^{t_0}, N\Gamma_{t_0}(\zeta))$ with $\Gamma_{t_0}(\cdot)$ defined in (8). From **(AC1)** we have $K\rho^{t_0} \leq 1/4$, and from (8) and $K \geq 2$ we deduce that there exists $A > 0$ such that $\sup_{\|\zeta\| \geq A} \Gamma(\zeta) \leq 1/4$. Using $n = (t - v)/t_0$ with $v \in [0, t_0)$ and (21), we derive the expected inequality in Lemma 4.3. \square

4.2 Proof of Theorem 2.2

The density $f_{k,t}$ of the a.c. part of the probability distribution of $t^{-1/2}Y_t$ under \mathbb{P}_k is given by (see (14))

$$\forall y \in \mathcal{E}_t := t^{-1/2}\mathcal{D}_t, \quad f_{k,t}(y) = t^{d/2} g_{k,t}(t^{1/2}y).$$

Denote the closure and the frontier of \mathcal{E}_t by $\overline{\mathcal{E}}_t$ and $\partial\mathcal{E}_t$ respectively. The following properties of $f_{k,t}$ are easily deduced from **(AC2)**.

Lemma 4.4 *Condition **(AC2)** is assumed to hold. Then, for each $k = 1, \dots, N$ and for all $t > 0$, the function $f_{k,t}$ is continuous on $\overline{\mathcal{E}}_t$ and differentiable on \mathcal{E}_t . In addition, we have*

$$\sup_{y \in \overline{\mathcal{E}}_t} |f_{k,t}(y)| = O(t^{d/2}) \quad \sup_{y \in \partial\mathcal{E}_t} |f_{k,t}(y)| = O(t^{-1/2}) \quad \max_{1 \leq j \leq d} \sup_{y \in \mathcal{E}_t} \left| \frac{\partial f_{k,t}}{\partial y_j}(y) \right| = O(t^{(d+1)/2}).$$

For any $t > 0$, set $d(y, \partial\mathcal{E}_t) := \inf\{d(y, z), z \in \partial\mathcal{E}_t\}$ and define

$$\mathcal{E}'_t := \{y \in \mathcal{E}_t, d(y, \partial\mathcal{E}_t) > t^{-(d/2+1)}\}, \quad (25)$$

Theorem 2.2 follows from the two following propositions.

Proposition 4.1 *Under Condition **(AC2)**, there is a positive constant C such that*

$$\forall t \in \mathbb{T}, \forall y \in \mathbb{R}^d \setminus \mathcal{E}'_t, \quad |f_{k,t}(y) - \eta_I(y)| \leq C(t^{-1/2} + \sup_{y \notin \mathcal{E}_t} \eta_I(y)).$$

Proof. First let $y \in \mathcal{E}_t \setminus \mathcal{E}'_t$. By definition of $\mathcal{E}'(t)$, there exists some $y_t \in \partial\mathcal{E}_t$ such that $\|y - y_t\| \leq 2t^{-(d/2+1)}$. From Taylor's inequality and Lemma 4.4 it follows that

$$\begin{aligned} |f_{k,t}(y) - \eta_I(y)| &\leq |f_{k,t}(y) - f_{k,t}(y_t)| + |f_{k,t}(y_t) - \eta_I(y_t)| + |\eta_I(y_t) - \eta_I(y)| \\ &\leq 2t^{-(d/2+1)} O(t^{(d+1)/2}) + O(t^{-1/2}) + \sup\{\eta_I(x) : x \in \partial\mathcal{E}_t\} + 2t^{-(d/2+1)} \\ &\leq O(t^{-1/2}) + \sup\{\eta_I(x) : x \in \partial\mathcal{E}_t\}. \end{aligned}$$

Second, let $y \in \overline{\mathcal{E}}_t \setminus \mathcal{E}_t = \partial\mathcal{E}_t$. Then $|f_{k,t}(y) - \eta_I(y)| \leq O(t^{-1/2}) + \sup\{\eta_I(x) : x \in \partial\mathcal{E}_t\}$ from Lemma 4.4. Third, let $y \in \mathbb{R}^d \setminus \overline{\mathcal{E}}_t$. Then $|f_{k,t}(y) - \eta_I(y)| = |\eta_I(y)| \leq \sup\{\eta_I(x) : x \notin \mathcal{E}_t\}$. The proof of Proposition 4.1 is complete. \square

Proposition 4.2 *Assume that Conditions **(I-A)**, **(M3)**, and **(AC1)**-**(AC2)** hold true. Then there exists a positive constant C such that*

$$\forall t \in \mathbb{T}, \forall y \in \mathcal{E}'_t, \quad |f_{k,t}(y) - \eta_I(y)| \leq C t^{-1/2}.$$

Note that the inequality above is valid for $k \in \mathbb{X}$ and $t \in [0, t_0]$ where t_0 is given in **(AC1)** since η_I is bounded on \mathbb{R}^d and $f_{k,t}$ is uniformly bounded on \mathbb{R}^d with respect to k and $t \in [0, t_0]$ from Lemma 4.4.

Proof. For any $p \in \mathbb{N}^*$, let $\eta_{p,I}$ be the function defined by $\forall y \in \mathbb{R}^d, \eta_{p,I}(y) := p^d \eta_I(py)$. The usual convolution product on \mathbb{R}^d is denoted by the symbol \star . Write for all $p \in \mathbb{N}^*$:

$$\begin{aligned} |f_{k,t}(y) - \eta_I(y)| &\leq |f_{k,t}(y) - (\eta_{p,I} \star f_{k,t})(y)| + |(\eta_{p,I} \star f_{k,t})(y) - (\eta_{p,I} \star \eta_I)(y)| + |(\eta_{p,I} \star \eta_I)(y) - \eta_I(y)| \\ &:= K_{1,p}(y) + K_{2,p}(y) + K_{3,p}(y). \end{aligned} \quad (26)$$

The conclusion of Proposition 4.2 follows from the two next lemmas. \square

For $t > t_0$ we define the following positive integer $p_t := \lfloor t^{d+3/2} \rfloor + 1$.

Lemma 4.5 *There exists $D > 0$ such that*

$$\forall t \in [t_0, +\infty), \forall p \geq p_t, \forall y \in \mathcal{E}'_t, \quad K_{1,p}(y) + K_{3,p}(y) \leq D t^{-1/2}.$$

Lemma 4.6 *There exists $E > 0$ such that*

$$\forall t \in [t_0, +\infty), \forall y \in \mathbb{R}^d, \quad K_{2,p_t}(y) \leq E t^{-1/2}.$$

Proof of Lemma 4.5. Note that $\int_{\mathbb{R}} \eta_{p,I}(u) du = 1$. Let $y \in \mathcal{E}'_t$. Using the definition of \mathcal{E}'_t (see (25)) and the fact that \mathcal{E}_t is open, one can prove that, for all $u \in \mathbb{R}^d$ such that $\|u\| \leq t^{-(d/2+1)}$, we have $y + u \in \overline{\mathcal{E}}_t$ (prove that $\sup\{a \in [0, 1] : y + au \in \mathcal{E}_t\} = 1$). By applying Taylor's inequality to $f_{k,t}$ and using Lemma 4.4, we obtain

$$\begin{aligned} |f_{k,t}(y) - (\eta_{p,I} \star f_{k,t})(y)| &\leq \int_{\|u\| \leq t^{-(d/2+1)}} \eta_{p,I}(u) |f_{k,t}(y) - f_{k,t}(y - u)| du \\ &\quad + \int_{\|u\| > t^{-(d/2+1)}} \eta_{p,I}(u) |f_{k,t}(y) - f_{k,t}(y - u)| du \\ &\leq O(t^{-1/2}) + O(t^{d/2}) \int_{\|v\| > p t^{-(d/2+1)}} \eta_I(v) dv. \end{aligned}$$

Next, if $p \geq t^{d+3/2}$, then Markov's inequality gives

$$\int_{\|v\| > p t^{-(d/2+1)}} \eta_I(v) dv \leq (2\pi)^{-d/2} \int_{\|v\| > t^{(d+1)/2}} e^{-\|v\|^2/2} dv = O(t^{-(d+1)/2}).$$

This gives the claimed conclusion for $K_{1,p}(y)$. Similarly we can prove that $K_{3,p}(y) = O(t^{-1/2})$ using the fact that $\eta_I(\cdot)$ and its differential are bounded on \mathbb{R}^d . \square

Proof of Lemma 4.6. Since $f_{k,t}(y) = t^{d/2} g_{k,t}(t^{1/2}y)$, we have $\widehat{f}_{k,t}(\zeta) = \widehat{g}_{k,t}(t^{-1/2}\zeta)$. For any $t > t_0$ and $\zeta \in \mathbb{R}^d$, we set

$$\Delta_{k,t}(\zeta) := |\widehat{g}_{k,t}(t^{-1/2}\zeta) - e^{-\|\zeta\|^2/2}|.$$

Let $0 < \delta < A$ (fixed) be given by Lemmas 4.1 and 4.3. From $0 < \widehat{\eta}_{p,I}(\zeta) = \widehat{\eta}_I(p^{-1}\zeta) \leq 1$ and the inverse Fourier formula, the following inequality holds for all $p \in \mathbb{N}^*$ and $y \in \mathbb{R}^d$:

$$\begin{aligned} (2\pi)^d K_{2,p}(y) &\leq \int_{\mathbb{R}} \Delta_{k,t}(\zeta) \widehat{\eta}_I(p^{-1}\zeta) d\zeta \\ &\leq \int_{\|\zeta\| \leq \delta\sqrt{t}} \Delta_{k,t}(\zeta) d\zeta + \int_{\delta\sqrt{t} \leq \|\zeta\| \leq A\sqrt{t}} \Delta_{k,t}(\zeta) d\zeta + \int_{\|\zeta\| \geq A\sqrt{t}} \Delta_{k,t}(\zeta) \widehat{\eta}_I(p^{-1}\zeta) d\zeta \\ &:= J_1(t) + J_2(t) + J_{3,p}(t). \end{aligned}$$

From (14) we obtain $\phi_{k,t}(\zeta) = \mathbb{E}_k[e^{i\zeta Y_t}] = \widehat{g}_{k,t}(\zeta) + \widehat{\mu}_{k,t}(\zeta)$. Using this equality and (7), we obtain for all $t > t_0$:

$$\begin{aligned} J_1(t) &\leq \int_{\|\zeta\| \leq \delta\sqrt{t}} |\phi_{k,t}(t^{-1/2}\zeta) - e^{-\|\zeta\|^2/2}| d\zeta + c\rho^t O(t^{d/2}) := I_1(t) + O(t^{-1/2}). \\ J_2(t) &\leq \int_{\delta\sqrt{t} \leq \|\zeta\| \leq A\sqrt{t}} |\phi_{k,t}(t^{-1/2}\zeta) - e^{-\|\zeta\|^2/2}| d\zeta + c\rho^t O(t^{d/2}) := I_2(t) + O(t^{-1/2}). \end{aligned}$$

Now we obtain from Lemma 4.1 and Lemma 4.2 respectively

$$\begin{aligned}
I_1(t) &\leq \int_{\|\zeta\| \leq \delta\sqrt{t}} \left| \lambda(t^{-1/2}\zeta)^{[t]} L_{k,t}(t^{-1/2}\zeta) + R_{k,t}(t^{-1/2}\zeta) - e^{-\|\zeta\|^2/2} \right| d\zeta \\
&\leq \int_{\|\zeta\| \leq \delta\sqrt{t}} \left(|L_{k,t}(t^{-1/2}\zeta)| |\lambda(t^{-1/2}\zeta)^{[t]} - e^{-\|\zeta\|^2/2}| \right. \\
&\quad \left. + e^{-\|\zeta\|^2/2} |L_{k,t}(t^{-1/2}\zeta) - 1| + |R_{k,t}(t^{-1/2}\zeta)| \right) d\zeta \\
&\leq t^{-1/2} C \left[(1 + \delta C) \int_{\mathbb{R}^d} (1 + \|\zeta\|^3) e^{-\|\zeta\|^2/8} d\zeta + \int_{\mathbb{R}^d} \|\zeta\| e^{-\|\zeta\|^2/2} d\zeta \right] + C r^{[t]} \int_{\|\zeta\| \leq \delta\sqrt{t}} d\zeta \\
&\leq O(t^{-1/2}) + O(t^{d/2} r^{[t]}); \\
I_2(t) &\leq \int_{\delta\sqrt{t} \leq \|\zeta\| \leq A\sqrt{t}} |\phi_{k,t}(t^{-1/2}\zeta)| d\zeta + \int_{\delta\sqrt{t} \leq \|\zeta\| \leq A\sqrt{t}} e^{-\|\zeta\|^2/2} d\zeta \\
&\leq O(t^{d/2} \tau^{[t]}) + \int_{\|\zeta\| \geq \delta\sqrt{t}} e^{-\|\zeta\|^2/2} d\zeta.
\end{aligned}$$

Thus, we have proved that there exists a constant $e_1 > 0$ such that

$$\forall t > t_0, \quad J_1(t) + J_2(t) \leq e_1 t^{-1/2}. \quad (27)$$

It remains to prove that: $\forall t > t_0, J_{3,p_t}(t) = O(t^{-1/2})$. We have for any $p \geq 1$

$$\begin{aligned}
J_{3,p}(t) &= \int_{\|\zeta\| \geq A\sqrt{t}} |\widehat{g}_{k,t}(t^{-1/2}\zeta) - e^{-\|\zeta\|^2/2}| \widehat{\eta}_I(\zeta/p) d\zeta \\
&\leq \int_{\|\zeta\| \geq A\sqrt{t}} |\widehat{g}_{k,t}(t^{-1/2}\zeta)| \widehat{\eta}_I(p^{-1}\zeta) d\zeta + \int_{\|\zeta\| \geq A\sqrt{t}} e^{-\|\zeta\|^2/2} d\zeta
\end{aligned}$$

The last integral is $O(t^{-1/2})$. Next, picking $p = p_t$, we deduce from Lemma 4.3 that: $\forall t > t_0$

$$\int_{\|\zeta\| \geq A\sqrt{t}} |\widehat{g}_{k,t}(t^{-1/2}\zeta)| \widehat{\eta}_I(p_t^{-1}\zeta) d\zeta \leq NC \frac{t}{2^{t/t_0}} \int_{\mathbb{R}^d} \widehat{\eta}_I(p_t^{-1}\zeta) d\zeta \leq NC \frac{t p_t^d}{2^{t/t_0}} \int_{\mathbb{R}^d} \widehat{\eta}_I(\zeta) d\zeta.$$

Thus there exists a constant $e_2 > 0$ such that: $\forall t \geq t_0, J_{3,p_t}(t) \leq e_2 t^{-1/2}$. \square

Remark 4.1 *The uniform version of Theorem 2.2 in Remark 2.3 holds because the conclusions of Lemmas 4.1, 4.2 and 4.3 are valid uniformly in $P \in \mathcal{P}_0$ under (U1)-(U3). This is obvious for Lemma 4.3. The uniformity in $P \in \mathcal{P}_0$ for Lemmas 4.1 and 4.2 is derived from the fact that the map $(P, \zeta) \mapsto \widehat{\mathcal{Y}}_1(\zeta)$ is continuous from $\mathcal{P}_0 \times \mathbb{R}^d$ into $\mathcal{M}_N(\mathbb{C})$ under (U1) when the moment condition (M3) holds uniformly in $P \in \mathcal{P}_0$. See Section 6 in [HL] for details.*

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